

The DOE SciDAC Institute for Scalable Data Management, Analysis, and Visualization (SDAV)

Rob Ross Deputy Director SciDAC Institute for Scalable Data Management, Analysis, and Visualization

Computer Scientist

Mathematics and Computer Science Division

Argonne National Laboratory

rross@mcs.anl.gov





Outline

- SDAV Introduction and Overview
- HPC Architectures and Data [Management|Visualization|Analysis]
- Example SDAV Activities
- Wrap-Up





- DOE Office of Science (SC) activity, spanning multiple Office of Science programs
- Computer Science and Applied Mathematics (SciDAC Institutes) Provide expertise
 and software tools in applied mathematics and computer science to advance
 scientific discovery through modeling and simulation
 - FASTMath Frameworks, Algorithms, and Scalable Technologies for Mathematics
 - QUEST Quantification of Uncertainty in Extreme Scale Computations
 - SUPER Institute for Sustained Performance, Energy and Resilience
 - SDAV Scalable Data Management, Analysis and Visualization
- Applications (SciDAC Partnerships) Partner with SC programs to combine CS and applied math with domain science expertise to target areas of strategic importance
 - Fusion Plasma Science (2 projects)
 - High Energy Physics (3 projects)
 - Nuclear Physics (3 projects)
 - Earth Systems (3 projects)
 - Chemistry and Materials (6 projects)

http://www.scidac.gov/

R. Laviolette and C. Susut. SciDAC Scientific Computation Application Partnerships Update. ASCAC. August 14, 2012.



Goal is to assist application scientists in using state-of-the-art data management, analysis, and visualization techniques to make new science discoveries:

- Data Management infrastructure that captures the data models used in science codes, efficiently moves, indexes, and compresses this data, enables query of scientific datasets, and provides the underpinnings of in situ data analysis
- Data Analysis application-driven, architecture-aware techniques for performing in situ data analysis, filtering, and reduction to optimize downstream I/O and prepare for in-depth post-processing analysis and visualization
- Data Visualization exploratory visualization techniques that support understanding ensembles of results, methods of quantifying uncertainty, and identifying and understanding features in multi- scale, multi-physics datasets
- Funded by the DOE Office of Science Advanced Scientific Computing Research Program
- Lead by Arie Shoshani (LBNL)
- Focus is on users of largest DOE/ASCR computational resources
- http://www.sdav-scidac.org

The SDAV Team

Institute Director:

Arie Shoshani, LBNL

Deputy Director:

Robert Ross, ANL

Executive Council:

Arie Shoshani, LBNL (chair) James Ahrens, LANL Wes Bethel, LBNL Hank Childs, LBNL Scott Klasky, ORNL Kwan-Liu Ma, UC Davis Valerio Pascucci, U Utah Robert Ross, ANL

Key Personnel:

Sean Ahern, ORNL James Ahrens, LANL Wes Bethel, LBNL Peer-Timo Bremer, LLNL Eric Brugger, LLNL Phil Carns, ANL Hank Childs, LBNL Alok Choudhary, Northwestern Berk Geveci, Kitware Charles Hansen, U Utah Chris Johnson, U Utah Kenneth Joy, UC Davis Scott Klasky, ORNL Robert Latham, ANL Kwan-Liu Ma, UC Davis

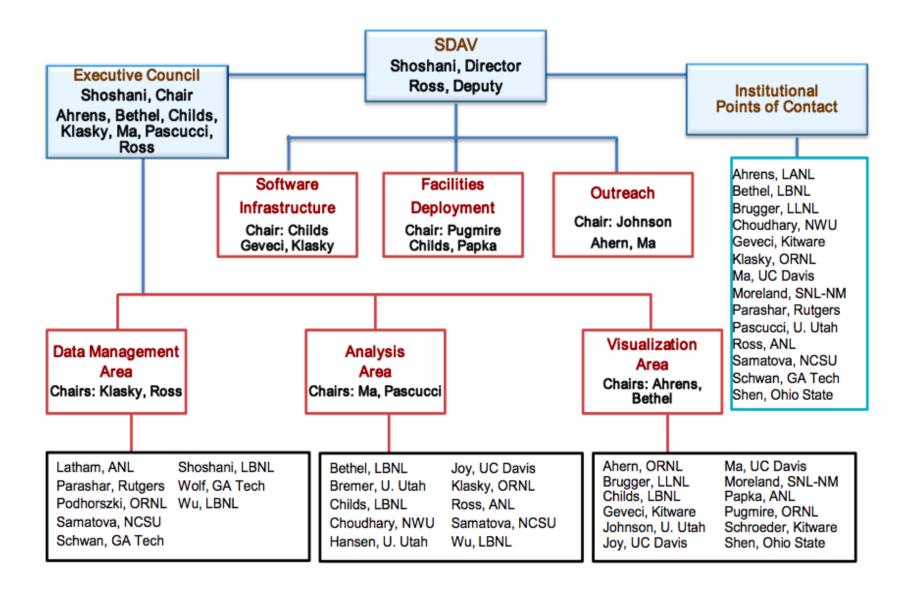
Kenneth Moreland, SNL Michael Papka, ANL Manish Parashar, Rutgers Valerio Pascucci, U Utah Tom Peterka, ANL Norbert Podhorszki, ORNL David Pugmire, ORNL Nagiza Samatova, NC State William Schroeder, Kitware Karsten Schwan, GA Tech Han-Wei Shen, OSU Venkat Vishwanath, ANL

Matthew Wolf, GA Tech Jonathan Woodring, LANL John Wu, LBNL

Six laboratories, seven universities, one company = diverse skill set.

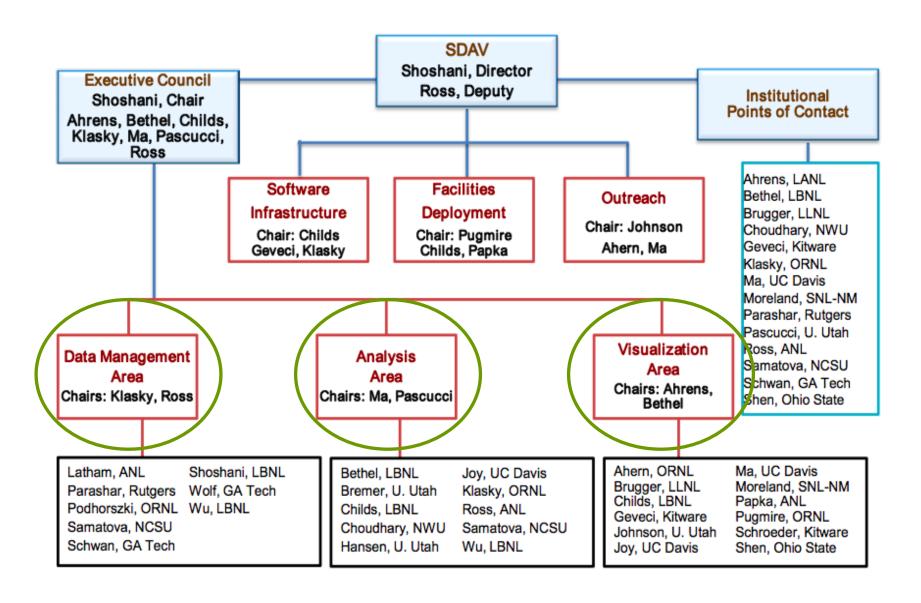


SDAV Organization





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The SDAV Institute Tools

Data Management tools

- In Situ Processing and Code Coupling
 - ADIOS
 - Glean
- Indexing
 - FastBit
- In Situ Data Compression
 - ISABELLA
- Parallel I/O and File Formats
 - PnetCDF
 - BP-files
 - (HDF5)
- I/O Characterization
 - Darshan

Data Analysis tools

- Statistical and Data Mining Techniques
 - NU-Minebench
- Importance-Driven Analysis Techniques
 - Domain-Knowledge Directed
 - Geometry Based
- Topological Methods
 - In Situ Topology
 - Feature-Based Analysis
 - High-Dimensional Analysis
- Building Blocks
 - DIY

Visualization tools

- Parallel visualization
 - Visit
 - ParaView
- VTK-m framework
- Flow Visualization Methods
- Rendering
- Ensembles, Uncertainty, and Higher-Dimensional Methods



SDAV Technology Use in Leadership Applications

Application	Code	Contact	Allocation (M node hours)	SDAV Technologies
Astrophysics	Chimera	T. Mezzacappa	60	ADIOS, VisIt, Ultravis-V
Astrophysics	FLASH	D. Lamb	80	PnetCDF, GLEAN, ROMIO, VisIt, VTK
Astrophysics	Maestro	J. Bell	50	VisIt
Astrophysics	Enzo	M. Norman	35	ParaView, VisIt
Biology	Nektar	G. Karniadakis	50	ParaView
Climate	POP	P. Jones	110	PnetCDF, ParaView, ROMIO
Combustion	S3D	J. Chen	60	ADIOS, Dataspaces, Ultravis-V, Ultravis-P, ViSUS IDX, Topologika
Combustion	Boxlib	J. Bell	60	VisIt, ADIOS, Topologika
Combustion	Nek5000	C. Frouzakis	150	VisIt
Cosmology	HACC	S. Habib	150	ParaView, ROMIO, Ultravis-P
Fusion	GTC	Z. Lin	35	ADIOS, DataTap, FastBit, Ultravis-V
Fusion	XGC	C.S. Chang	50	ADIOS, Dataspaces, FastBit, Ultravis-V, VTK
Fusion	GTC-P	W. Tang	58	ADIOS, Ultravis-V, Ultravis-P
Plasma	VPIC	B. Daughton	30	PnetCDF, ParaView, ROMIO
Nuclear	Nek5000	P. Fischer	25	ROMIO, Visit



Data and HPC Architectures

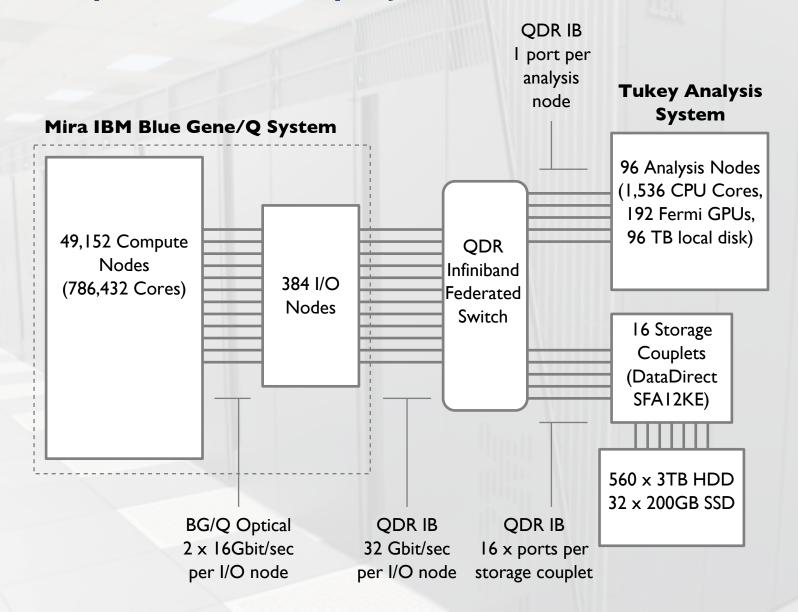


Exascale Systems: Potential Architecture

Systems	2009	2018*	Difference
System Peak	2 Pflop/sec	1 Eflop/sec	O(1000)
Power	6 Mwatt	20 Mwatt	
System Memory	0.3 Pbytes	32-64 Pbytes	O(100)
Node Compute	125 Gflop/sec	1-15 Tflop/sec	O(10-100)
Node Memory BW	25 Gbytes/sec	2-4 Tbytes/sec	O(100)
Node Concurrency	12	O(1-10K)	O(100-1000)
Total Node Interconnect BW	3.5 Gbytes/sec	200-400 Gbytes/sec	O(100)
System Size (Nodes)	18,700	O(100,000-1M)	O(10-100)
Total Concurrency	225,000	O(1 billion)	O(10,000)
Storage	15 Pbytes	500-1000 Pbytes	O(10-100)
1/0	0.2 Tbytes/sec	60 Tbytes/sec	O(100)
MTTI	Days	O(1 day)	

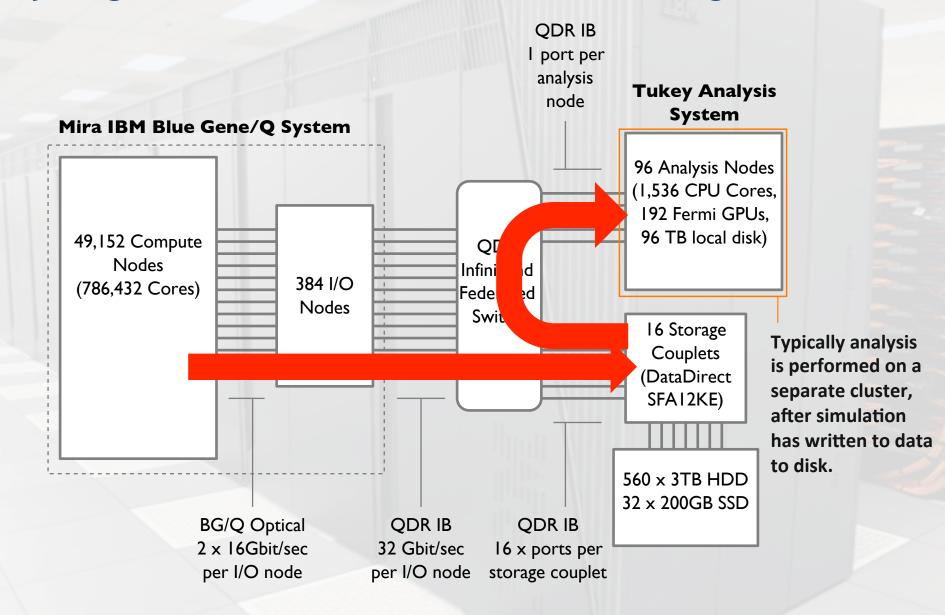
From J. Dongarra, "Impact of Architecture and Technology for Extreme Scale on Software and Algorithm Design," Cross-cutting Technologies for Computing at the Exascale, February 2-5, 2010.

An Example Leadership System Architecture



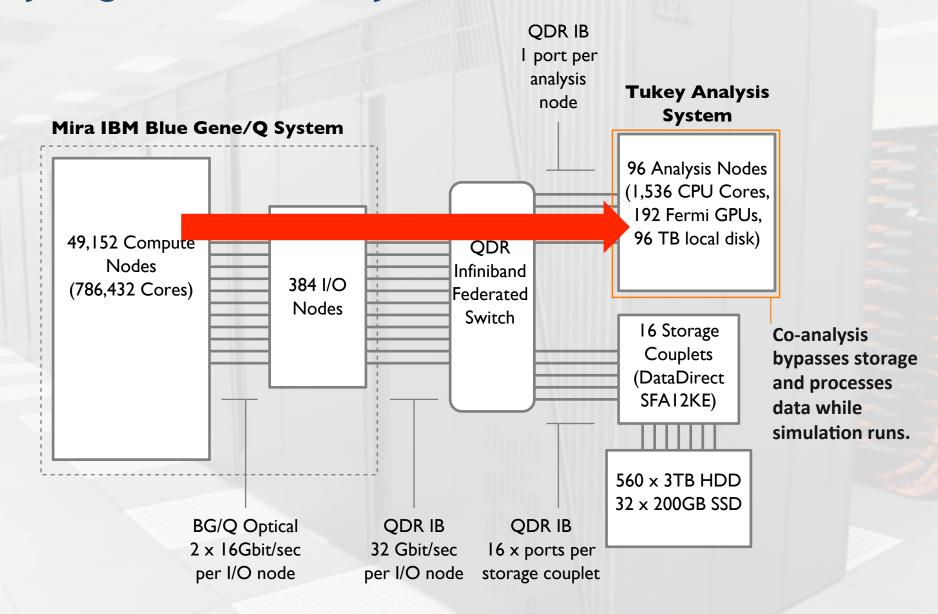


Analyzing Data: Traditional Post-Processing



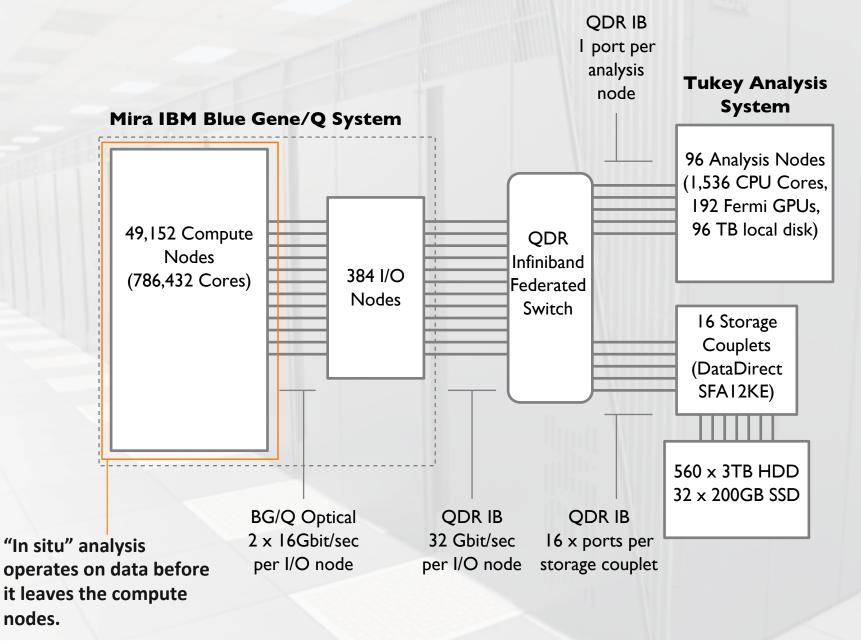


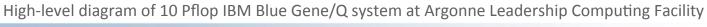
Analyzing Data: Co-Analysis





Analyzing Data: In Situ Analysis





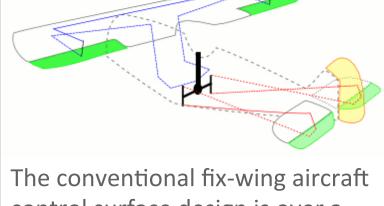
nodes.

A Few Examples



Improving Aircraft Designs

- Goal is to reduce fuel consumption, noise, and drag in commercial aircraft:
 - Redesigning the vertical tail of a commercial jet could reduce jet fuel use by 0.5%, resulting in annual savings of \$300 million.
 - One new aircraft control method employs synthetic jets; understanding the behavior (e.g., frequency, amplitude, location) is critical for future aircraft wing design
- Synthetic jet simulations are conducted using the PHASTA CFD solver and are being correlated with experimental data



The conventional fix-wing aircraft control surface design is over a century old. New control methods are being developed that improve efficiency and enable new aircraft designs.

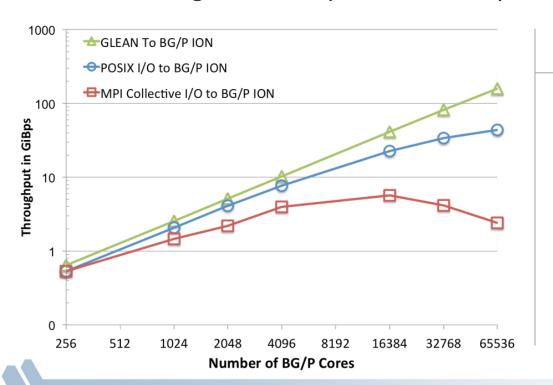
Image by Piotr Jaworski, released under GNU Free Documentation License.

- Collaboration between Univ. of Colorado, Boulder, Rensselaer Polytechnic Institute, and Boeing
- Adaptive unstructured mesh code, has scaled to 4.3 billion mesh elements and 160K cores on ALCF Blue Gene/P



Streamlining Data Movement in Airflow Simulation

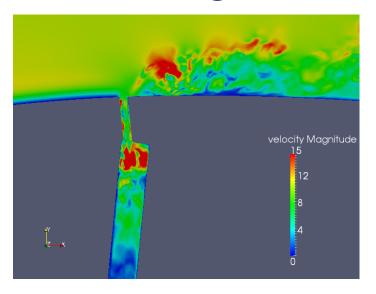
- PHASTA CFD simulations produce as much as ~200 GB per time step
 - Rate of data movement off compute nodes determines how much data the scientists are able to analyze
- GLEAN is a flexible and extensible framework for simulation-time data movement and analysis
 - Accelerating I/O via topology awareness, asynchronous I/O
 - Enabling in situ analysis and co-analysis

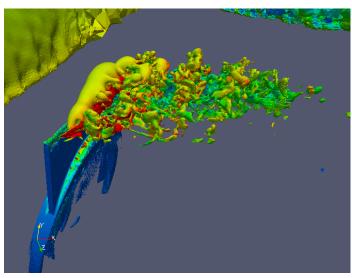


Strong scaling performance for 1GB data movement off ALCF Intrepid Blue Gene/P compute nodes. GLEAN provides 30-fold improvement over POSIX I/O at large scale. Strong scaling is critical as we move towards systems with increased core counts.

Thanks to V. Vishwanath (ANL) for providing this material.

Observing Simulated Synthetic Jet Behavior





Cut plane through synthetic jet (left) and isosurface of vertical velocity (right) colored by velocity (both for 3.3 billion element mesh). Analysis performed with ParaView.

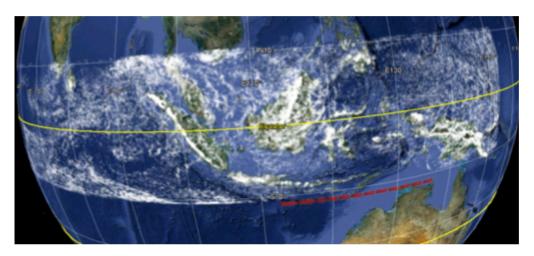
Thanks to K. Jansen (UC Boulder) for these images.

- Using GLEAN, scientists are able to use co-analysis to observe simulation behavior at run time and avoid storage bottlenecks
 - In co-analysis, data is moved from compute to analysis resources without first being stored on disk
 - Reduces storage requirements, overlaps analysis with simulation, and achieves very data throughput (48 GiBps)
- This enables the scientists to better understand the temporal characteristics of the synthetic jet
 - Cost of analyzing a timestep is much lower, so scientists can view results at a higher temporal fidelity than was feasible before (approx. every 10 timesteps)

Understanding the Madden-Julian Oscillation (MJO)

- MJO is a 30-60 day oscillation of enhanced and suppressed rainfall near the Indian and western Pacific Oceans
 - MJO can be thought of as a wavefront indicating how the cloud system is moving
 - Understanding the phenomenon helps explain tropical weather variations
 - Also related to summer precipitation patterns in North America
- Simulation performed by R. Leung and S. Hagos (PNNL) using ARW-WRF3.1
 - 2700 x 600 x 27 (vertical) curvilinear mesh
 - 480 timesteps representing 120 days (multiple iterations of phenomenon)
 - 3GB per timestep

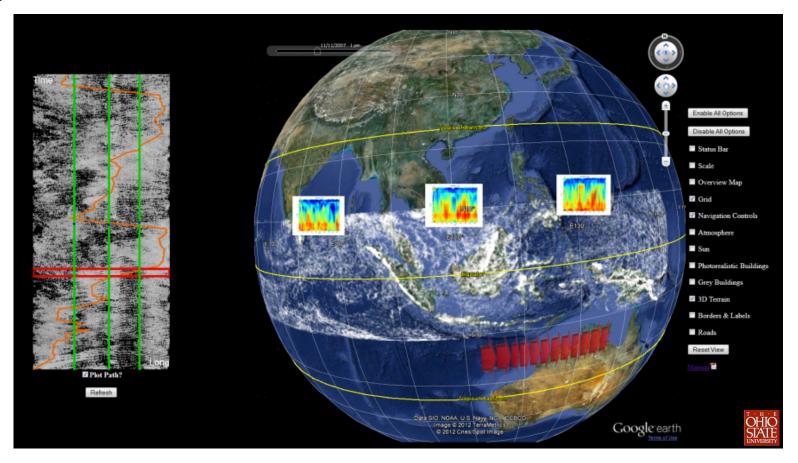
The image shows a rendering of clouds on a virtual globe — interface developed by SDAV researchers at Ohio State University, in collaboration with P. C. Wong, S. Hagos, and R. Leung (PNNL).



S. Hagos, L. R. Leung, and J. Dudhia. Thermodynamics of the Madden-Julian oscillation in a regional model with constrained moisture. Journal of Atmospheric Sciences, 68:1974–1989, 2011.

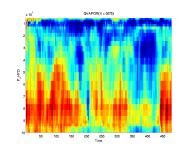
Interactively Exploring the MJO Phenomenon

SDAV members H. Shen and T. Lee (OSU), with collaborator P.C. Wong (PNNL) developed an interface to assist scientists in identifying and exploring the MJO phenomenon in simulations.

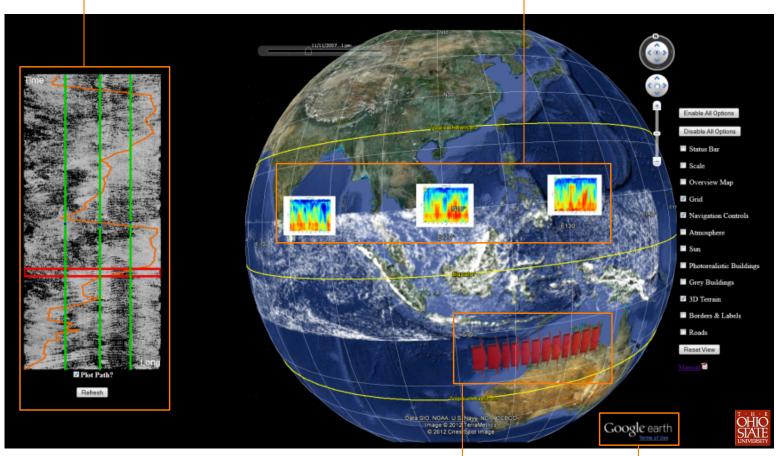


A typical method of viewing water vapor mixing ratio is via a timeline view, with time on the Y axis and longitude on the X axis (known as a Hovmoller diagram).

Orange line tracks highest water vapor mixing ratio over time (MJO path), while red box selects an interval of time for visualization on the right. Green lines show heatmap locations.



Heatmap views are generated on demand when user selects a longitude. Heatmap shows water vapor mixing ratio by altitude (Y axis) over time (X axis).



Thanks to H. Shen and T. Lee (OSU) for providing this material.

Red arrows indicate direction of cloud system movement in the selected time period.

Builds off the widely-available Google Earth platform, can be embedded in web pages.

Understanding How a Laser Pulse Propagates Through a Hydrogen Plasma

- VORPAL code used to simulate laser wakefield particle accelerator
 - 3D simulation
 - 30 timesteps
 - 90 million particles per timestep, ~5 Gbytes of data per timestep
- Questions:
 - Which particles become accelerated? How are they accelerated?
 - How did the beam form? How did it evolve?
- Data management, analysis, and visualization:
 - Data model support HDF5, H5Part to store data with appropriate metadata
 - Indexing FastBit to enable quick identification of particles of interest, associate particles between timesteps
 - Visualization Parallel coordinates view to help user select particles, VisIt as deployment vehicle



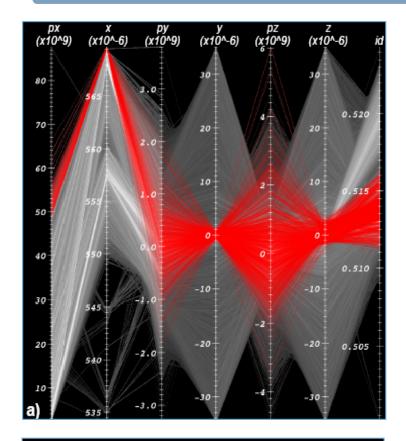
Beam Selection

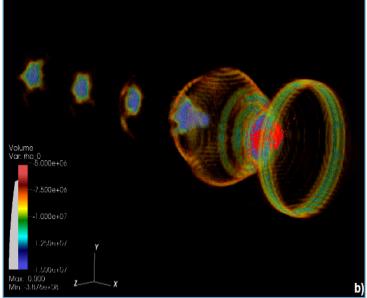
Parallel coordinates view of t = 12

- Grey particles represent initial selection $(px > 2*10^9)$
- Red particles represent "focus particles" in first wake period following pulse $(px > 4.856*10^{10}) \&\& (x > 5.649*10^{-4})$

Volume rendering of plasma density with focus particles included in red (t = 12)

Helps locate beam within wake





Thanks to E. Wes Bethel (LBNL) for providing this material.

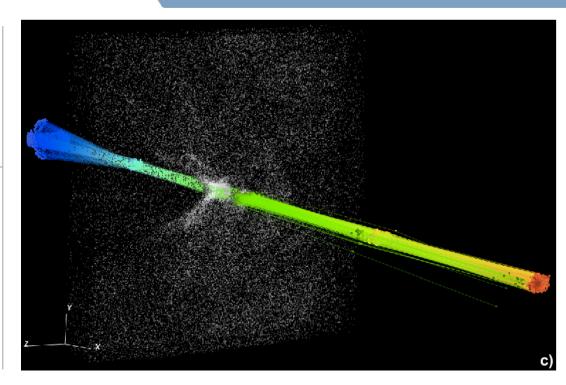
Tracing Particles over Time

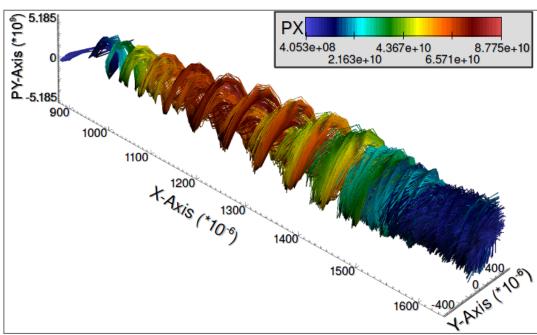
Tracing particles back to t = 9 and forward to t = 14 allows scientist to see acceleration over time:

- Heatmap shows particles constantly accelerated over time (increase in px, left to right).
- Grey particles show initial selection (for reference).

More recent work shows:

- Particles start out slow (blue, left), undergo acceleration (reds), then slow again as the plasma wave outruns them (blue, right).
- Spiral structure shows particles oscillating transversely in the focusing field (new science).





Thanks to E. Wes Bethel (LBNL) for providing this material.

Final Comments: Accomplishing Our Goal

A mix of activities contribute to the success:

- Community Engagement Actively engaging application teams running on leading DOE computing systems, our sibling Institutes, and DOE computing facility personnel over the lifetime of the Institute.
- Technology Deployment Working with application scientists so that they
 can use state of the art tools and techniques to support their needs in
 data management, analysis, and visualization tasks.
- Research Integration Incorporating ASCR basic research results into our portfolio and developing new technologies as needed to meet the needs of application scientists over the next five years.
- Software Support Performing quality software deployment, maintenance, and support to ensure the success of our tools.

Computational science applications are data intensive. SDAV is assisting scientists in using state-of-the-art tools and techniques to manage this data and glean new science discoveries.

SDAV Contacts

- Director: Arie Shoshani <<u>shoshani@lbl.gov</u>>
- Deputy Director: Me <<u>rross@mcs.anl.gov</u>>
- Data Management Leads
 - Scott Klasky <<u>klasky@ornl.gov</u>>
 - Me <<u>rross@mcs.anl.gov</u>>
- Analysis Leads
 - Kwan-Liu Ma <ma@cs.ucdavis.edu>
 - Valerio Pascucci <<u>pascucci@sci.utah.edu</u>>
- Visualization Leads
 - Jim Ahrens <<u>ahrens@lanl.gov</u>>
 - Wes Bethel <<u>ewbethel@lbl.gov</u>>